

Sensitivity/Uncertainty Analysis of a Borehole Scenario Comparing Latin Hypercube Sampling and Deterministic Sensitivity Approaches

Technical Report

October 1983

William V. Harper Sumant K. Gupta

Office of Nuclear Waste Isolation 505 King Avenue Battelle Memorial Institute Columbus, Ohio 43201



BATTELLE Project Management Division

BIBLIOGRAPHIC DATA

Harper, William V., and Sumant K. Gupta, 1983. Sensitivity/Uncertainty Analysis of a Borehole Scenario Comparing Latin Hypercube Sampling and Deterministic Sensitivity Approaches, BMI/ONWI-516, Office of Nuclear Waste Isolation, Battelle Memorial Institute, Columbus, OH.

NOTICE

This report was prepared as an account of work sponsored by an agency of the United States Government. Neither the United States Government nor any agency thereof, nor any of their employees, makes any warranty, expressed or implied, or assumes any legal liability or responsibility for the accuracy, completeness, or usefulness of any information, apparatus, product, or process disclosed, or represents that its use would not infringe privately owned rights. Reference herein to any specific commercial product, process, or service by trade name, trademark, manufacturer, or otherwise, does not necessarily constitute or imply its endorsement, recommendation, or favoring by the United States Government or any agency thereof. The views and opinions of authors expressed herein do not necessarily state or reflect those of the United States Government or any agency thereof.

Printed in the United States of America Available from National Technical Information Service U.S. Department of Commerce 5285 Port Royal Road Springfield, VA 22161

NTIS price codes
Printed copy: A03
Microfiche copy: A01

Sensitivity/Uncertainty Analysis of a Borehole Scenario Comparing Latin Hypercube Sampling and Deterministic Sensitivity Approaches

Technical Report

October 1983

William V. Harper Sumant K. Gupta

Office of Nuclear Waste Isolation 505 King Avenue Battelle Memorial Institute Columbus, Ohio 43201

This report was prepared by the Battelle Project Management Division, Office of Nuclear Waste Isolation under Contract No. DE-AC 02-83CH10140 with the U.S. Department of Energy.

ACKNOWLEDGMENTS

The authors would like to thank J. M. Furr, Office of Nuclear Waste Isolation, for his computer programming expertise, especially on the Latin Hypercube Sampling software. In addition, they would like to acknowledge the extremely useful conversations with E. M. Oblow and D. G. Cacuci, both of Oak Ridge National Laboratory, on adjoint theory and applications. An excellent review by J. L. Wilson, INTERA Environmental Consultants, has considerably improved this report. Also, the authors gratefully acknowledge the funding of this work by the U.S. Department of Energy.

ABSTRACT

A computer code was used to study steady-state flow for a hypothetical borehole scenario. The model consists of three coupled equations with only eight parameters and three dependent variables. This study focused on steady-state flow as the performance measure of interest. Two different approaches to sensitivity/uncertainty analysis were used on this code. One approach, based on Latin Hypercube Sampling (LHS), is a statistical sampling method, whereas, the second approach is based on the deterministic evaluation of sensitivities.

The LHS technique is easy to apply and should work well for codes with a moderate number of parameters. Of deterministic techniques, the direct method is preferred when there are many performance measures of interest and a moderate number of parameters. The adjoint method is recommended when there are a limited number of performance measures and an unlimited number of parameters. This unlimited number of parameters capability can be extremely useful for finite element or finite difference codes with a large number of grid blocks.

The Office of Nuclear Waste Isolation will use the technique most appropriate for an individual situation. For example, the adjoint method may be used to reduce the scope to a size that can be readily handled by a technique such as LHS. Other techniques for sensitivity/uncertainty analysis, e.g., kriging followed by conditional simulation, will be used also.



CONTENTS

1	INTRO	DDUCTI	ON			•	• •	•	• •	٠	•	٠	•	•	•	•	•	•	•	•	•	•	•	•	•	•	i
2	DEMON	NSTRAT	ION	SCE	ENAR	10		•				•		•	•		•	•					•				2
3	ANALY	YSIS T	ЕСН	NIQU	JES			•						•										•	•	•	6
	3.1	STATI	STI	CAL	SAM	PLI	NG	MET	HOD														•			•	6
		3.1.1	D	esig	gn M	latr	ix	Gen	era	tio	n					•					•		•	•		•	6
		3.1.2	2 B	orel	hole	Sc	ena	rio	An	aly	's i	S				•			•		•	•		•	•	•	8
		3.1.3	3 A	na l	ysis	Re	sul	ts			•	•		•				•	•			•		•	•	•	9
	3.2	DETER	RMIN	IST	IC E	VAL	UAT	ION	I ME	THC	DS			•				•	•	•	•	•	•	•	٠	•	10
		3.2.1	l M	lath	ema t	ica	1 D	eri	vat	ior	1			•	•	•			•	•	•		•	•	•	•	11
		3.2.2	2 R	Resu	lts	of	Det	ern	ini	sti	c	Ba	se	d	An	a 1	ys	es	•	•	•	•	•	•	•	•	14
4	COMP	ARATIV	/E A	NAL	YSIS	RE	SUL	.TS			•			•						•				•	•		18
	4.1	INITI	[AL	METI	HODS	CO	MPA	RIS	SON										•	•				•		•	18
	4.2	ADDIT	FION	IAL (COMF	ARI	SON	IS							•					٠.		•	•			•	20
	4.3	CUMUL	_AT I	(VE	DIST	RIB	UTI	ON	FUN	ICT I	ON	1							•			•	•		•	•	22
5	REFE	RENCES	ŝ.											•		•								•		•	25
ΑP	PENDI:	х A -	LA1	rin	HYPE	ERCU	BE	SAM	1PL I	NG	DE	SI	GN	M	ΙAΤ	RI	CE	S		•	•		•				27
ΑP	PENDI:	ХВ-	STE	PWI	SE F	REGR	ESS	101	١.										•	•				•			31
ΑP	PENDI	X C -	DET	ΓERM	INIS	STIC	SE	ENS!	ITIV	'IT	Y (:0E	FF	IC	ΙE	EN7	\$		•							•	33
AP	PENDI	X D -	RAI SI	NGE ENSI	OF I	DETE ITY	RM: COI	INIS EFF	STIO ICIE	C NO	ori S	1AI	.17	ΈC		•						•			•	•	37

•

LIST OF FIGURES

	<u>Pa</u>	ige
2-1.	Borehole Demonstration Schematic	2
2-2.	Borehole Scenario	4
4-1.	Empirical CDF for Q	23
4-2.	Empirical CCDF for Q	24
	LIST OF TABLES	
2-1.	Input Parameter Probability Distributions	5
3-1.	Key Parameters	9
3-2.	LHS Model 2 Regression Estimates	10
3-3.	Partial Derivatives Required for Deterministic Sensitivity Analysis	13
3-4.	Deterministic Sensitivities	15
3-5.	Estimated Performance Measure Uncertainty	17
4-1.	LHS Approach Analysis Results	18
4-2.	Deterministic Approach Analysis Results	19
4_3	Further Comparisons of Methods	21

			•

1 INTRODUCTION

The Office of Nuclear Waste Isolation (ONWI) will conduct extensive sensitivity/uncertainty studies as part of their performance assessment of a high-level nuclear waste repository in salt. The assessments will quantify performance measures and their related uncertainty. The sensitivity of the performance measures will be evaluated using computer models. These models focus primarily on continuous processes such as ground-water flow with discrete events (e.g., human intrusion) superimposed on the process behavior.

The purpose of this report is to provide a description of two sensitivity/uncertainty analysis approaches: (1) a statistical sampling method based on Latin Hypercube Sampling (LHS); and (2) a deterministic evaluation of sensitivities coupled with a parameter covariance matrix for uncertainty. Both the direct and adjoint deterministic methods are described. A hypothetical borehole flow scenario was used to demonstrate these analysis technologies. The types of information obtainable from the various techniques are described. Analysis results are provided and comparisons of results are presented.

2 DEMONSTRATION SCENARIO

A hypothetical scenario is described in which a borehole is drilled through an aquifer above a nuclear waste repository, through the repository, to an aquifer below. The flow through this borehole is a function of parameters such as the radius of the borehole, hydraulic potential difference between lower and upper aquifers, conductivity of fill material, transmissivity of upper and lower aquifers, etc. Each parameter varies considerably. For this scenario, eight input parameters are considered and analyses are made using a simple computer code.

Figure 2-1 provides a schematic view of a borehole drilled from the ground surface through a high-level nuclear waste repository and the aquifers above and below the repository horizon. As a conservative assumption, the borehole is considered as screened through the full depths of both aquifers. For this analysis, it is assumed that the potentiometric head is higher in the upper aquifer than in the lower. This results in downward flow. This condition is found in the bedded salt sites being considered for a potential nuclear waste repository.

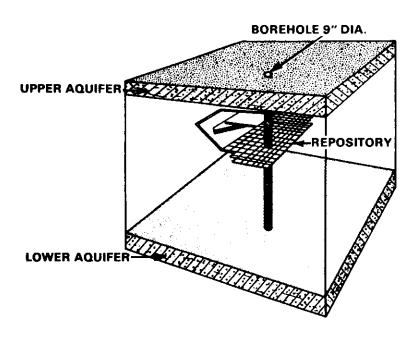


Figure 2-1. Borehole Demonstration Schematic.

Considering a fully penetrating well and no regional ground-water gradient, the steady-state flow through upper aquifer to borehole in an aquifer system is defined by Equation 1:

$$Q = \pi 2 T_u \frac{(H_u - H_{wu})}{\ln (r/r_w)}$$
 (1)

where:

 $Q = flow, m^3/yr$

 T_{μ} = transmissivity of upper aquifer, m^2/yr

 H_{u} = potentiometric head of upper aquifer, m

r = radius of influence, m

 r_w = radius of borehole, m

Using assumptions of Equation 1, the steady-state flow from borehole to lower aquifer can be described by Equation 2:

$$Q = -2\pi T_1 \frac{(H_1 - H_{w1})}{\ln (r/r_w)}$$
 (2)

where:

 T_1 = transmissivity of lower aquifer, m^2/yr

 H_1 = potentiometric head of lower aquifer, m

Hwl = steady-state potentiometric head in borehole at lower
aquifer, m

Using Darcy's equation, the steady-state, laminar, and isothermal flow flow of a homogeneous fluid through the borehole can be described by Equation 3:

$$Q = \pi r_{w}^{2} K_{w} \frac{(H_{wu} - H_{w1})}{L}$$
 (3)

where:

 K_{ω} = hydraulic conductivity of borehole, m/yr

L = length of borehole, m

Parameters Q, H_{wl} , and H_{wl} are dependent, but Q will be the performance measure of interest. Figure 2-2 illustrates these parameters.

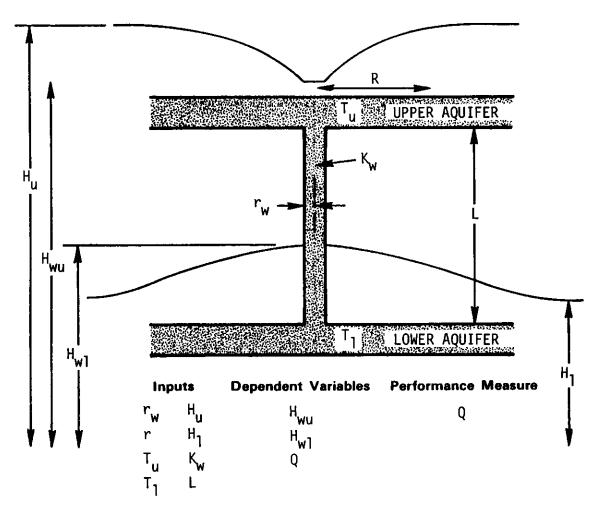


Figure 2-2. Borehole Scenario Parameters.

Attempts were made to define the probability distributions of the parameters using available information for typical salt sites. For those parameters where good statistical properties were not available, probability distributions were assumed and will be updated with the availability of additional data. At present, the distributions in Table 2-1 are assumed to be statistically independent implying zero pairwise covariance or correlation between the parameters. This assumption neglects the causal relationship of $K_{\mathbf{w}}$ to $\mathbf{r}_{\mathbf{w}}$ and of \mathbf{r} to transmissivity.

Table 2-1. Input Parameter Probability Distributions.

Input Parameter	Range	Distribution			
r _w	0.05 to 0.15 m	Normal ($\mu = 0.10$, $\sigma = 0.0161812$)*			
r	100 to 50,000 m	Lognormal ($\mu' = 7.71$, $\sigma' = 1.0056$)**			
T _u	63,070 to 111,600 m ² /yr	Uniform			
Hu	990 to 1,110 m	Uniform			
т,	63.1 to 116 m ² /yr	Uniform			
н ₁	700 to 820 m	Uniform			
L.	1,120 to 1,680 m	Uniform			
K,	9,855 to 12,045 m/yr	Uniform			

 $\overset{*}{\mu}_{\mu}$, σ are the mean and standard deviation, respectively, of r_{W} . $\overset{**}{\mu}_{\mu}$, σ' are the mean and standard deviation, respectively, of the ln (r) which is normally distributed.

Probability distributions of the input parameters in Table 2-1 were used to generate LHS input settings. They were also used to develop the covariance matrix needed for uncertainty analyses using deterministic sensitivity coefficients.

3 ANALYSIS TECHNIQUES

Two sensitivity/uncertainty analysis approaches are described to quantify the sensitivity of the performance measure (flow rate) to the input parameters and the contribution of each parameter to the overall uncertainty in flow rate. Latin Hypercube Sampling (LHS), as used by the U.S. Nuclear Regulatory Commission (NRC) in risk studies for nuclear waste repositories (Helton and Iman, 1980, and Pepping et al, 1983), and commonly associated statistical techniques are applied for the scenario analysis. The second approach is a deterministic evaluation of sensitivity coefficients plus statistical techniques to quantify the desired uncertainty.

As shown in this report, both approaches provide a suitable tool for sensitivity/uncertainty analysis of the hypothetical borehole scenario. It should be emphasized that, as in any analytic technique, no method is universally superior. The suitability of one method over others depends on the number of parameters, nonlinearity in the system, the size of the situation being analyzed, and the magnitude of uncertainty in the parameters.

3.1 STATISTICAL SAMPLING METHOD

The statistical sampling method described herein is based on the LHS technique. LHS is a member of a family of statistical sampling techniques which includes Monte Carlo sampling and stratified random sampling. LHS has been used in many risk assessments performed by Sandia National Laboratories for the NRC (Helton and Iman, 1980; Iman and Conover, 1980a,b; and Pepping et al, 1983).

3.1.1 Design Matrix Generation

Latin Hypercube Sampling is used to generate what is called a design matrix. Specifically, if N computer runs are to be made of the computer code to be analyzed with k parameters under study, the design matrix will be N x k. Each row of the matrix contains the input settings for each of the k parameters being studied. As explained in ONWI-444 (Harper, 1983), LHS has several potential advantages over the other sampling techniques (McKay et al, 1979) in that it covers the range of each input parameter and provides

a method of introducing a desired rank correlation matrix for these parameters (Iman and Conover, 1980a).

Parameter probability density functions and any rank correlation between parameters must be specified so that this design matrix can be generated. Typically, for an N x k design matrix, each of the k parameters is divided into N equi-probable intervals. Then a random sample is chosen from each interval in a manner that preserves the individual probability density function. In this manner, k N-tuples (an N x l vector) of input settings are determined. If the parameters are all assumed to be statistically independent (implying the covariance matrix of the parameter is a diacratic matrix), then these N-tuples are randomly joined together to form the N x k design matrix. If independence of these parameters is not a reasonable assumption, then the desired rank correlation matrix specified is used to match the N-tuples in such a manner that approximates this correlation structure.

After the design matrix is established by LHS, other statistical techniques are used to establish the sensitivities and related uncertainty. It should be emphasized that LHS strictly ends with the development of the design matrix. In statistical parlance, it creates the experimental design. Since the design created is done so at random, it does not have nice properties such as orthogonality of classical experimental designs developed for bypothesis testing. (Classical screening designs usually sample only a few points in the range of values for each parameter (Hicks, 1973) as opposed to LHS covering the entire spectrum more fully.) Commonly, stepwise regression analysis and/or partial correlation are used in the identification of the key parameters, i.e., the most sensitive parameters.

The design matrix for N computer runs is used as input to the computer code of interest and the desired outputs (performance measures) are recorded for each of these N runs. The performance measures of interest can be studied as a function of the input parameters using techniques such as stepwise regression.

Often, the analysis is performed on the ranks of the data if highly nonlinear relationships are present between a performance measure and the inputs. The rank transformation replaces the raw data values by the

integers 1, 2, ..., N if no ties exist between the values. If ties are present, the average rank is used for the tied values.

3.1.2 Borehole Scenario Analysis

Two sample sets of inputs were generated using LHS (Iman et al, 1980): one with 10 design points and the other with 50 design points. Each design point represents a separate combination of the eight parameters. The two design matrices created by LHS (N = 10 and N = 50) are shown in Appendix A.

The N = 50 design matrix was used as input to the borehole scenario flow computer code. For each of the 50 randomly generated design points, the computer code was run creating an output file that contained the design point and the corresponding value of performance measure Q. This output data set was then input into the MINITAB (Ryan et al, 1981) statistical software package. Various descriptive and inferential statistical techniques were used in the analysis of these data. The major technique utilized was stepwise regression. A summarized procedure for the stepwise regression technique is provided as Appendix B.

The parameter combination H_U - H_I was used as additional input to the stepwise regression analyses performed. Two models were evaluated. Model 1 is shown as Equation 4:

$$\hat{Q} = b_0 + \sum_{i} b_i x_i + \sum_{i} b_i j x_i x_j$$
 (Model 1)

where:

 \hat{Q} = predicted flow, m^3/yr

b = coefficient(s) of Model 1

x = parameters (eight initial plus H_u - H₁)

Thus, quadratic and all cross-product terms are considered. Model 2 (linear terms only) is shown as Equation 5:

$$\hat{Q} = a_0 + \sum_{i} a_i x_i$$
 (Model 2)

where a = the coefficient(s) of Model 2. (Stepwise regression was performed considering both models for the raw data and the ranks of the data. Stepwise

regression on the ranks did not provide a better fit than the raw data, so only the raw data were considered for the remainder of the analyses.)

Least squares estimates of coefficients (b or a) were obtained for the variables identified as significant in the stepwise regression procedure. The variability of the performance measure $\mathbb Q$ is divided into two parts. The first is the variability accounted for by the regression equation (Eq. 4 or 5), whereas, the remainder is the "residual" variability not covered by the regression equation. One measure of the "goodness of fit" of the regression is $\mathbb R^2$, the percentage of variability of performance measure $\mathbb Q$ accounted for by the regression equation (or 100 times the ratio of the first part to the total variability of $\mathbb Q$).

3.1.3 Analysis Results

Key parameters identified by this statistical sampling method, and the associated R^2 are shown (Table 3-1). The combination of variables given are parsimonious and are the "best" predictors for performance measure Q. Both models provide equally good fits to the data and involve the same variables.

Table 3-1. Key Parameters.

Model 1 (allows quadratic and cross-product terms)	Model 2 (linear terms only)
r _w (H _u -H ₁)	r _w
(H _u -H ₁)L	H _u -H ₁
K _w	L
	K _W
$R^2 = 98.0$	R ² = 97.6

Table 3-2 presents estimated coefficients and the percent variability accounted for (from \mathbb{R}^2) for the four parameters of Model 2. Model 2 will be used in the remainder of this report.

Estimates.								
Parameter	Estimated Coefficient	% Variance of Q						
r _w	1405.0	70.5						
Hu-Hl	0.229	15.6						

-0.054

0.005

10.2

1.3

Table 3-2. LHS Model 2 Regression Estimates.

3.2 DETERMINISTIC EVALUATION METHODS

Statistical methods for sensitivity analysis, such as those presented based on LHS, typically fit a polynomial to approximate the relationship between the desired output and the inputs. However, this relationship is explicitly detailed in the computer code itself. Deterministic approaches to sensitivity analysis take advantage of this fact. The two major deterministic approaches are the direct method and the adjoint method. The direct method is more appropriate when there are few parameters under study and many performance measures, whereas, the adjoint approach is superior when there are a limited number of performance measures and many parameters.

Obtaining parameter sensitivities using these deterministic methods is usually much more complicated than using statistical techniques. It requires differentiating the mathematical relationships used with respect to the performance measure of interest. ONWI-380 (Thomas, 1982), ONWI-444 (Harper, 1983), Oblow, 1978a,b, or Cacuci et al, 1982, describe in detail the process involved in deterministic evaluation methods. The mathematics used to obtain deterministic sensitivities for a simple borehole flow scenario is described. Partial derivatives and normalized sensitivity coefficients are obtained.

3.2.1 Mathematical Derivation

The steps given below show how deterministic sensitivity methods are applied to the borehole flow scenario analysis. Both the direct and adjoint methods are presented. The three governing equations (Eq. 1, 2, and 3) may be rewritten as follows:

$$Q = \frac{2\pi T_u (H_u - H_{wu})}{\ln (r/r_w)} = S_1 (H_u - H_{wu})$$
 (1')

$$Q = -\frac{2\pi T_1 (H_1 - H_{w1})}{\ln (r/r_w)} = S_2 (H_1 - H_{w1})$$
 (2')

$$Q = \frac{\pi r_{w}^{2} K_{w} (H_{wu} - H_{w1})}{L} = S_{3} (H_{wu} - H_{w1})$$
 (3')

The response or performance measure of interest is flow, Q, with $H_{\rm wl}$ and $H_{\rm wu}$ as additional dependent variables and the rest as independent parameters. It is now necessary to differentiate Q with respect to a generic parameter, α , which can be any of the eight input parameters.

$$\frac{\partial Q}{\partial \alpha} = \frac{\partial S_1}{\partial \alpha} \left(H_u - H_{wu} \right) + S_1 \left(\frac{\partial H_u}{\partial \alpha} - \frac{\partial H_{wu}}{\partial \alpha} \right)$$
 (6)

$$\frac{\partial Q}{\partial \alpha} = \frac{\partial S_2}{\partial \alpha} \left(H_1 - H_{w1} \right) + S_2 \left(\frac{\partial H_1}{\partial \alpha} - \frac{\partial H_{w1}}{\partial \alpha} \right)$$
 (7)

$$\frac{\partial Q}{\partial \alpha} = \frac{\partial S_3}{\partial \alpha} \left(H_{wu} - H_{w1} \right) + S_3 \left(\frac{\partial H_{wu}}{\partial \alpha} - \frac{\partial H_{w1}}{\partial \alpha} \right)$$
 (8)

Letting $h_1 = \partial Q/\partial \alpha$, $h_2 = \partial H_{w1}/\partial \alpha$, and $h_3 = \partial H_{wu}/\partial \alpha$, and rearranging terms, we have Equations 9 through 11:

$$h_1 + S_1 h_3 = \frac{\partial S_1}{\partial \alpha} (H_u - H_{wu}) + S_1 \frac{\partial H_u}{\partial \alpha}$$
 (9)

$$h_1 + S_2 h_2 = \frac{\partial S_2}{\partial \alpha} (H_1 - H_{w1}) + S_2 \frac{\partial H_{w1}}{\partial \alpha}$$
 (10)

$$h_1 + S_3 h_2 - S_3 h_3 = \frac{\partial S_3}{\partial \alpha} (H_{wu} - H_{w1})$$
 (11)

or, Equation 12:

$$\begin{pmatrix} 1 & 0 & S_1 \\ 1 & S_2 & 0 \\ 1 & S_3 & -S_3 \end{pmatrix} \begin{pmatrix} h_1 \\ h_2 \\ h_3 \end{pmatrix} = \begin{pmatrix} \frac{\partial S_1}{\partial \alpha} (H_u - H_{wu}) + S_1 \frac{\partial H_u}{\partial \alpha} \\ \frac{\partial S_2}{\partial \alpha} (H_1 - H_{w1}) + S_2 \frac{\partial H_1}{\partial \alpha} \\ \frac{\partial S_3}{\partial \alpha} (H_{wu} - H_{w1}) \end{pmatrix}$$

$$A \qquad h \qquad S$$

$$(12)$$

At this point, we could solve for h as follows:

$$Ah = S \rightarrow h = A^{-1} S \tag{13}$$

(This is the direct method and results in partial derivatives for each of the three dependent variables, Q, H_{wu} , and H_{wl} , i.e., $h_1 = \partial Q/\partial \alpha$, $h_2 = \partial Q/\partial \alpha$, and $h_3 = \partial Q/\partial \alpha$.)

To solve for the desired sensitivities, partial derivatives for S_1 , S_2 , and S_3 must be specified for each α (the eight input parameters). These partial derivatives are provided in Table 3-3.

Table 3-3. Partial Derivatives Required for Deterministic Sensitivity Analysis.

		•	
Ct.	∂S ₁ /∂α	∂S ₂ /∂α	∂S ₃ /∂α
T _u	2π 1n(r/r _w)	0	0
H _u	0	0	0
r	-	2π ^T 1 r[ln(r/r _W)] ²	0
r _w	2πT _u [ln(r/r _w)] ² rr _w	-2πT ₁ [ln(r/r _w)] ² rr _w	2πr _w K _w /L
т ₁	0	-2π ^T] r[ln(r/r _W)] ²	0
н	0	0	0
K _W	0	0	πr _w ² /L -πr _w ² K _w /L ²
L	0	0	-πr _w ² K _w /L ²

The derivation of adjoint sensitivities is detailed. The response of interest will be denoted by R. Thus, R = Q, and we want to find the partial derivatives $\partial R/\partial \alpha$. Let g'=(1,0,0) where the prime indicates a transpose. Then $g'h=h_1=\partial Q/\partial \alpha=\partial R/\partial \alpha$. Given Eq. 12, multiply by an arbitrary row vector, v', to obtain

$$v'(Ah-S) = 0 (14)$$

Then $\partial R/\partial \alpha$ may be rewritten as

$$\frac{\partial R}{\partial \alpha} = g'h - v' \text{ (Ah-S)}$$
 (15)

or

$$\frac{\partial R}{\partial \alpha} = (g' - v'A)h + v'S \qquad (16)$$

To remove h, set (g'-v'A) to zero and solve for v:

$$g'-v'A = 0 \rightarrow v = (A')^{-1} g$$
 (17)

Now the desired partials may be found from Equation 15 as

$$\frac{\partial R}{\partial \alpha} = v'S \tag{18}$$

Notice that we now may find $\partial R/\partial \alpha$ without needing to know the vector h. The vector S will change for each α using Table 4.

3.2.2 Results of Deterministic Based Analyses

The borehole flow program was modified to include the steps presented earlier for the development of direct and adjoint approaches. The results presented in Table 3-4 were based on a design point that was felt to be representative of the physical system. For the seven input parameters

with symmetric distributions, the midpoint or mean of the range was selected. For r with a lognormal distribution, the geometric mean of the range was chosen. For this combination of values, the first partial derivatives $(\partial Q/\partial \alpha)$ were derived as well as normalized sensitivity coefficients. Normalized sensitivity coefficients, $(\partial Q/\partial \alpha)(\alpha/Q)$, show the predicted percentage change in the performance measure, Q, for a given percentage change in α .

Parameter	Units	Value	9 Q/ ∂α	(3Q/3a)(a/Q)
T _u	m ² /yr	89,340.0	0.35 x 10 ⁻⁵	0.44 x 10 ⁻⁵
Hu	m	1,050.0	0.245	3.62
r	m	2,236.0	-0.14×10^{-4}	-0.0004
rw	m	0.1	1412.9	1.99
т ₁	m ² /yr	89.55	0.003	0.004
н ₁	m	760.0	-0.245	-2.62
K _w	m/yr	10,950.0	0.006	0.996
Ļ	m	1,400.0	-0.05	-0.996

Table 3-4. Deterministic Sensitivities.

Since H_u and H_l by themselves are meaningless in this scenario, they can be combined into H_u-H_l . The partial, $\partial Q/\partial \alpha$, for $\alpha=H_u-H_l$ can be found from the following:

0.245
$$H_u - 0.245 H_1 = 0.245 (H_u - H_1) \rightarrow \frac{\partial Q}{\partial (H_u - H_1)} = 0.245$$
 (19)

The normalized sensitivity coefficient is found by adding the normalized sensitivity coefficient for ${\rm H}_{\rm u}$ and ${\rm H}_{\rm l}$. Thus,

$$\frac{\partial Q}{\partial (H_u - H_1)} \cdot \frac{H_u - H_1}{Q} = 3.62 - 2.62 = 1.00$$
 (20)

It can be shown, in general, that the normalized sensitivity coefficient for $H_u^-H_l^-$ will always be largardless of the individual values of H_u^- and H_l^- . In a similar fashion, the normalized sensitivity coefficients of K_w^- and A will always add to 0. In many situations, the normalized sensitivity coefficients are more useful than the first partials of the performance measure because one is often interested in the effect of percent changes in the parameters rather than absolute changes. For example, a selected percentage change in r_w^- would cause roughly twice that percentage change in Q. Appendix D shows the stability of the normalized sensitivity coefficients over the ranges given in Table 2-1.

It is necessary to combine the partial derivatives with the parameter covariance matrix to estimate the uncertainty of the performance measure about the design point. This covariance matrix includes the parameter variance and correlation information and is combined with the partials as follows:

$$\hat{\sigma}_{\text{pm}}^2 = \S \Sigma \S' \tag{21}$$

where:

 σ_{pm}^2 = uncertainty or variance estimate of the performance measure

\$ = vector containing the first partial derivatives of performance
measure

 Σ = parameter covariance matrix.

For our scenario, Σ has a very simple structure due to the assumed independence of the parameters, i.e., Σ is a diagonal matrix with parameter variances down the diagonal. Thus

$$\hat{\sigma}_{Q}^{2} = \sum_{i} \sum_{j} \hat{\sigma}_{i}^{2}$$
 (22)

where:

 s_i = element i of s_i σ_i^2 = variance of parameter i Carrying out the calculations of Eq. (21) and (22) and determining the percent of the variance accounted for by each parameter results are shown in Table 3-5 where H_u - H_1 is used in place of H_u and H_1 separately.

Table 3-5. Estimated Performance Measure Uncertainty.

Parameter	% Uncertainty
r _w	71.6
r	
T _u	
H _u -H ₁	17.9
T ₁	
L	8.5
K _w	2.0

4 COMPARATIVE ANALYSIS RESULTS

4.1 INITIAL METHODS COMPARISON

At this point, it is worthwhile to compare the results obtained by the LHS approach and the deterministic approach. This initial comparison is based on the N = 50 sample for LHS and the 1 design point for the deterministic. Tables 4-1 and 4-2 summarize the results of this comparison.

Parameter	Units	Stepwise Regression Ranking	Regression Coefficient	% Uncertainty
r _w	m	1	1,405.0	70.5
r	m			
T _u	m ² /yr			
H _u -H ₁	m	2	0.229	15.6
т ₁	m ² /yr			
L	m	3	-0.054	10.2
K _W	m/yr	4	0.005	1.3

Table 4-1. LHS Approach Analysis Results.

A comparison of Table 4-1 and 4-2 finds very similar results for both approaches. Deterministic sensitivity usually uses the absolute value of the normalized sensitivity coefficients to rank the parameters in order of importance. Doing this results in $r_{\rm w}$ being ranked first, $H_{\rm u}-H_{\rm l}$ second, with L and $K_{\rm w}$ tied for third. This closely parallels the results of the LHS approach. The regression coefficient estimates obtained by the LHS approach are very close to the first partial derivatives obtained by the direct and adjoint techniques. Also, the percent uncertainty for the parameters are in close agreement for the two methods.

Table 4-2. Deterministic Approach Analysis Results.								
Parameter	Units	Normalized Sensitivity Coefficient	First Partial Derivative	% Uncertainty				
r _w	m	1.99	1412.9	71.6				
r	m	-0.00044	-0.14 x 10 ⁻⁴					
T _u	m ² /yr	0.000004	0.35 x 10 ⁻⁸					
H _u -H ₁	m	1.0	0.245	17.9				
T ₁	m ² /yr	0.0044	0.003					
L	m	-0.996	-0.05	8.5				
K _W	m/yr	0.996	0.006	2.0				

Table 4-2. Deterministic Approach Analysis Results.

Summarizing this, we see very similar results for both techniques on this relatively simple computer code to model flow through a borehole. The LHS approach is much simpler to apply for one familiar with standard statistical techniques; however, the direct and adjoint results presented used only one run of the computer code to find all the sensitivities. If quick turnaround is needed on a new code and there are a moderate number of parameters, LHS can be used to identify the key parameters. For a large problem with thousands of parameters, the adjoint approach provides a comprehensive screening of all parameters that is not possible with statistical techniques alone. However, analytic solution of the adjoint equations may take many person-months to complete.

As explained on ONWI-444 (Harper, 1983), ONWI needs capabilities such as the adjoint technique to provide a comprehensive screening of all parameters for licensing. To expedite the time it takes to develop an adjoint version of the code, Oak Ridge National Laboratory will develop a compiler called GRESS (Oblow, 1983a). The Gradient-Enhanced Software System (GRESS) uses computer calculus (Oblow, 1983b) on a FORTRAN-based computer code to greatly reduce the time and cost of

developing adjoint codes. As GRESS evolves, deterministic sensitivity procedures can be used in situations where one cannot afford to wait a year or longer for the necessary mathematics to be performed analytically on complex codes.

The major part of the initial comparisons used the 50 design point set for the LHS approach; whereas 1 design point was used for the deterministic methods. However, the close agreement found betweem the methods moved us to consider further comparisons.

4.2 ADDITIONAL COMPARISONS

The LHS and deterministic methods on both the N = 10 and N = 50 design matrices generated by LHS are compared (Table 4-3). This is done strictly to compare the sensitivity coefficients obtained by the two methods. In practice, one would not run the adjoint code for 50 design points; however, multiple design points may be needed for highly nonlinear codes in which the parameters exhibit high variability. These two design matrices and the corresponding dependent variable values may be found in Appendix A. Appendix C gives the deterministic sensitivity results for the N = 10 case.

Closer agreement is found between the N = 50 comparison of deterministic and LHS sensitivity coefficients than the N = 10 comparison though both are within reasonable agreement. For the N = 10 case, K_W was not found to be statistically significant by stepwise regression. Since a model with linear terms only for the independent parameters provided a good fit to the data using regression techniques, this comparison between the two approaches makes sense. When quadratic and cross-product terms enter the model, such a simple comparison is not feasible. In that case, when one changes the value of a parameter, one must consider the impact of any quadratic and cross-product terms when predicting the new value of the performance measure.

Table 4-3. Further Comparisons of Methods.

Table 1 of 1 at one of the chods.					
		N = 50		N = 10	
Parameter	Units	LHS Regression Coefficient	Average Deterministic Partial Derivative	LHS Regression Coefficient	Average Deterministic Partial Derivative
r _w	m	1405.0	1434.0	1484.0	1440.0
, r	m		-0.287 x 10 ⁻⁴		-0.259 x 10 ⁻⁴
T_{u}	m ² /yr		0.458 x 10 ⁻⁸		0.435×10^{-8}
H _u -H ₁	m	0.229	0.255	0.216	0.252
Ťl	m ² /yr		0.005		0.004
L	m	-0.054	-0.054	-0.06	-0.054
K _w	m/yr	0.005	0.007		0.007

4.3 CUMULATIVE DISTRIBUTION FUNCTION

When using the LHS approach, it is possible to develop an empirical probability distribution for the performance measure of interest. If the assumed input parameter distributions and correlations are correct, and the computer code is a valid representation of the relationship between the inputs and outputs, then the empirical probability distribution should provide a good estimate of the true probability distribution of the performance measure. A cumulative distribution function (CDF) often provides more useful risk information. Estimates of the probability of being below a specific flow rate for Q may be read directly off the empirical CDF seen in Figure 4-1. Some individuals prefer using a complementary cumulative distribution function (CCDF) as seen in Figure 4-2. This allows one to directly obtain probability estimates of exceeding a certain flow rate rather than having to compute it from the CDF.

To obtain representative CDF and CCDF, it is necessary to use a Monte Carlo sampling method such as LHS to determine the input settings. While deterministic methods may be coupled with sampling methods (as described in Section 4.2), usually not enough computer runs are performed to develop good empirical CDF. This is a plus from an efficiency point of view for sensitivity analysis; however, it is generally insufficient for the establishment of an empirical CDF. If an emperical CDF was an important part of the analysis being performed, one must make sure to couple it with a sampling approach to get the design points and to perform a sufficient number of computer runs to result in a reasonable estimate of the true CDF.

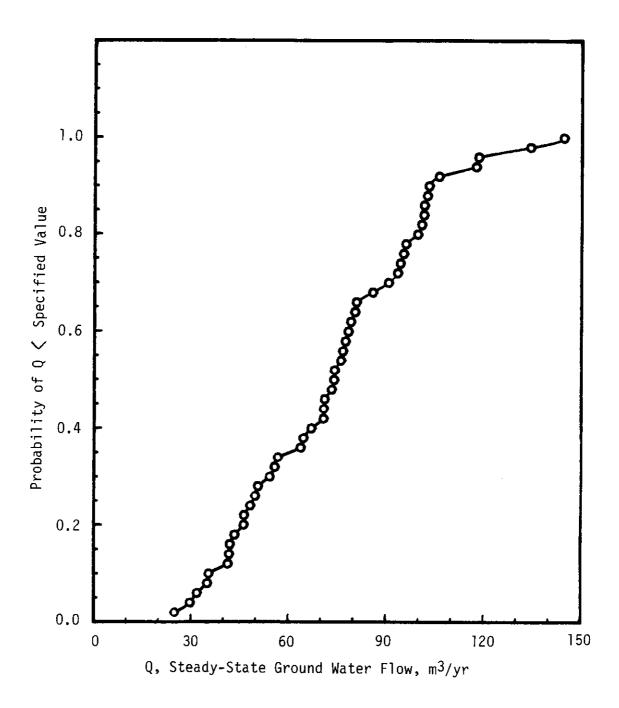


Figure 4-1. Empirical Cumulative Distribution Function for Q.

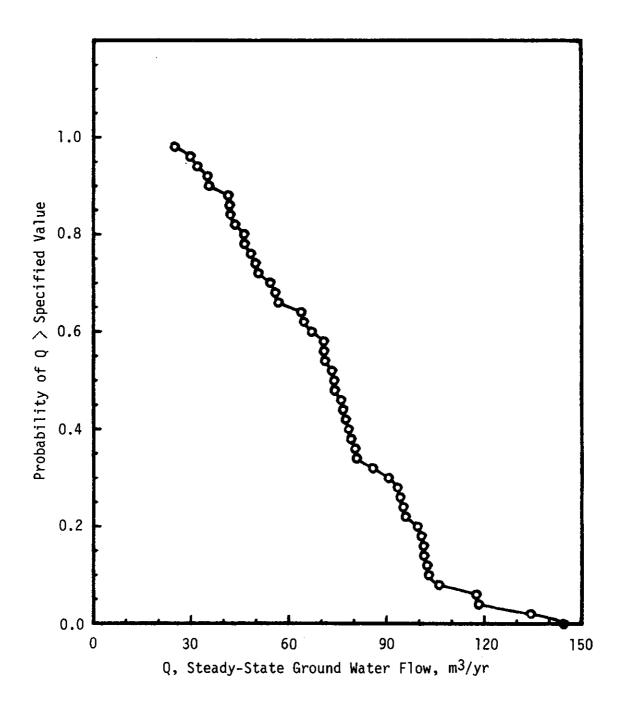


Figure 4-2. Empirical Complementary Cumulative Distribution Function for Q.

5 REFERENCES

- Cacuci, D. G., and E. Wacholder, "Adjoint Sensitivity Analysis for Transient Two-Phase Flow," <u>Nuclear Science and Engineering</u>, December, 1982.
- Harper, W. V., 1983. <u>Sensitivity/Uncertainty Analysis Techniques for Nonstochastic Computer Codes</u>, ONWI-444, Office of Nuclear Waste Isolation, Battelle Memorial Institute, Columbus, OH
- Hicks, C. R., 1973. <u>Fundamental Concepts in the Design of Experiments</u>, Hole, Rinehart, and Winston, New York, NY.
- Helton, J. C., and R. L. Iman, 1980. <u>Risk Methodology for Geologic Disposal of Radioactive Waste: Sensitivity Analysis of the Environmental Transport Model</u>, NUREG/CR-1636, Vol. 2, SAND79-1393, AN, Sandia National Laboratories, Albuquerque, NM.
- Iman, R. L., and W. J. Conover, 1980a. <u>Risk Methodology for Geologic Disposal of Radioactive Waste: A Distribution-Free Approach to Inducing Rank Correlation Among Input Variables for Simulation Studies</u>, SAND80-0157, Sandia National Laboratories, Albuquerque, NM.
- Iman, R. L., and W. J. Conover, 1980b. "Small Sample Sensitivity Analysis Techniques for Computer Models, With an Application to Risk Assessment," <u>Communications in Statistics</u>, A9, 17, pp. 1749-1874.
- Iman, R. L., J. M. Davenport, and D. K. Zeigler, 1980. <u>Latin Hypercube Sampling (Program User's Guide)</u>, SAND79-1473, Sandia National Laboratories, Albuquerque, NM.
- McKay, M. D., R. J. Beckman and W. J. Conover, 1979. "A Comparison of Three Methods for Selecting Values of Input Variables in the Analysis of Output from a Computer Code," <u>Technometrics</u>, Vol. 21, pp. 239-245.
- Oblow, E. M., 1978a. "Sensitivity Theory for Reactor Thermal-Hydraulics Problems," Nuclear Science and Engineering, Vol. 68, pp. 322-337.
- Oblow, E. M., 1978b. "Sensitivity Theory for General Nonlinear Algebraic Equations with Constraints," <u>Nuclear Science and Engineering</u>, Vol. 65, pp. 187-191.

Oblow, E. M., 1983a. <u>GRESS, Gradient-Enhanced Software System, Version B,</u> User's Guide, ORNL/TM-8339, Oak Ridge National Laboratory, Oak Ridge, TN.

Oblow, E. M., 1983b. An Automated Procedure for Sensitivity Analysis Using Computer Calculus, ORNL/TM-8776, Oak Ridge National Laboratory, Oak Ridge, TN.

Pepping, R. E., M. S. Y. Chu, K. K. Wahi, and N. R. Ortiz, 1983. <u>Risk</u>
<u>Analysis Methodology for Spend Fuel Repositories in Bedded Salt: Final Report,</u>
SAND81-2409, NUREG/CR-2402, Sandia National Laboratories, Albuquerque, NM.

Ryan, T. A., Jr., B. L. Joiner, and B. F. Ryan, 1981. MINITAB Reference Manual, Statistics Department, Pennsylvania State University, University Park, PA.

Thomas, R. E., 1982. <u>Uncertainty Analysis</u>, ONWI-380, prepared by Battelle Columbus Laboratories for Office of Nuclear Waste Isolation, Battelle Memorial Institute, Columbus, OH.

APPENDIX A - LATIN HYPERCUBE SAMPLING DESIGN MATRICES

In this appendix, the Latin Hypercube Sampling (LHS) design matrices are presented (Table A-1 and A-2) for the two cases discussed, N = 10 and N = 50. In addition, the three dependent variables (Q, H_{WU} , and H_{W1}) have their resultant values given for each of the design points. The variables are presented in the column order indicated below:

rw r Tu Hu T H I L Kw	m m m ² /yr m m ² /yr m m/yr	Values generated by LHS
Q H _{wu} H _{w1}	m ³ /yr m	Values generated by running borehole flow computer code for each of N design points

Table A-1. LHS Boreflow Results, N = 10.

Run No.	r _w	r	T _u	H _u	Tl	H ₁	L	K _w	H _{w1}	H _{wu}	Q
1	0.8609E-01	2948.0	0.8337E+05	1044.0	107.0	783.0	1250.0	0.1001E+05	783.754	1044.0	48.5247
2	0.1050	5194.0	0.8840E+05	1093.0	67.7	788.0	1181.0	0.1126E+05	790.538	1093.0	99.8815
3	0.1180	1358.0	0.9471E+05	993.0	98.6	758.0	1466.0	0.1174E+05	759.236	992.999	81.8886
4	0.9050E-01	240.0	0.1091E+06	1037.0	81.2	811.0	1534.0	0.1052E+05	811.614	1037.0	39.7706
5	0.9287E-01	1861.0	0.9219E+05	1101.1	103.0	711.0	1575.0	0.1075E+05	712.101	1101.0	71.9224
6	0.1250	1165.0	0.1015E+06	1055.0	86.5	734.0	1325.0	0.1023E+05	736.033	1055.0	120.885
7	0.9778E-01	902.0	0.6798E+05	1072.0	76.9	715.0	1380.0	0.1116E+05	716.631	1072.0	86.3202
8	0.1100	2616.0	0.6944E+05	1085.0	113.0	799.0	1429.0	0.1147E+05	800.233	1085.0	86.8866
9	0.1000	14690.0	0.7427E+05	1009.0	70.8	768.0	1672.0	0.1043E+05	769.257	1009.0	46.9832
10	0.7457E-01	8017.0	0.1153E+06	1025.0	90.6	738.0	1141.0	0.1202E+05	739.071	1025.0	52.6204

Table A-2. LHS Boreflow Results, N = 50.

Run No.	r _w	r	T _u	H _u	T ₁	н	L	K _w	H _{w1}	H _{wu}	Q
1	0.1200	307.0	0.7133E+05	1070.0	87.4	704.0	1504.0	0.1080E+05	705.691	1070.0	118.347
2	0.1210	7787.0	0.8976E+05	1088.0	88.5	812.0	1672.0	0.1071E+05	813.610	1083.0	80.8426
3	0.1040	2684.0	0.7764E+05	1092.0	91.4	782.0	1245.0	0.1120E+05	783.667	1092.0	94.2502
4	0.7928E-01	455.0	0.6463E+05	1046.0	63.1	801.0	1556.0	0.1147E+05	801.776	1046.0	35.5482
5	0.1170	480.0	0.9364E+05	1034.0	64.2	730.0	1430.0	0.1129E+05	732.114	1034.0	102.499
6	0.1010	1508.0	0.1029E+06	1006.0	114.0	737.0	1590.0	0.1051E+05	737.762	1006.0	56.8220
7	0.1130	16040.0	0.1038E+06	1004.0	103.0	714.0	1352.0	0.1179E+05	715.848	1004.0	100.801
8	0.1120	1121.0	0.8194E+05	1036.0	105.0	702.0	1282.0	0.1008E+05	703.439	1036.0	103.045
9	0.1000	24370.0	0.1060E+06	1023.0	84.1	711.0	1480.0	0.1177E+05	712.819	1023.0	77.4957
10	0.8917E-01	3967.0	0.1044E+06	1063.0	82.1	750.0	1154.0	0.9943E+04	751.392	1063.0	67.0667
11	0.9504E-01	1798.0	0.7483E+05	997.0	95.3	753.0	1138.0	0.1067E+05	754.063	996.999	64.6362
12	0.7425E-01	4686.0	0.1081E+06	1017.0	104.0	778.0	1311.0	0.1014E+05	778.540	1017.0	31.9442
13	0.9985E-01	1985.0	0.7907E+05	1044.0	71.2	725.0	1211.0	0.1106E+05	727.006	1044.0	90.6787
14	0.9564E-01	750.0	0.9121E+05	1025.0	113.0	793.0	1544.0	0.1156E+05	793.629	1025.0	49.7792
15	0.1150	6951.0	0.9751E+05	1008.0	69.2	799.0	1523.0	0.9870E+04	800.415	1008.0	55.8928
16	0.9209E-01	11710.0	0.1019E+06	1109.0	80.9	723.0	1525.0	0.1198E+05	724.859	1109.0	80.3990
17	0.1100	274.0	0.1120E+06	1039.0	112.0	745.0	1257.0	1.1075E+05	746.058	1039.0	95.2331
18	0.7821E-01	3134.0	0.8075E+05	1009.0	73.7	734.0	1651.0	0.1100E+05	734.803	1009.0	35.1060
19	0.7562E-01	2346.0	0.1071E+06	993.0	81.3	810.0	1303.0	0.1185E+05	810.603	993.0	29.7998
20	0.1080	2134.0	0.1134E+06	1072.0	103.0	819.0	1392.0	0.1116E+05	820.131	1072.0	73.9938
21	0.1030	2869.0	0.7306E+05	1014.0	110.0	813.0	1623.0	0.1017E+05	813.620	1014.0	41.8485
22	0.1040	10170.0	0.8365E+05	1079.0	100.0	727.0	1453.0	0.1171E+05	728.754	1079.0	95.9132
23	0.8143E-01	1359.0	0.6610E+05	1053.0	86.9	747.0	1500.0	0.1141E+05	747.861	1053.0	48.3514
24	0.1300	6148.0	0.9541E+05	1058.0	95.7	781.0	1179.0	0.1086E+05	783.404	1058.0	134.290
25	0.9691E-01	840.0	0.9264E+05	992.0	75.2	816.0	1226.0	0.1204E+05	816.973	991.999	50.7137

Table A-2. LHS Boreflow Results, N = 50 (Continued).

Run No.	r _w	r	T _u	H _u	τ ₁	Н	L	Κ _W	H _w }	H _{wu}	Q
26	0.9251E-01	664.0	0.1009E+06	1099.0	90.2	785.0	1379.0	0.1166E+05	786.114	1099.0	71.1289
27	0.9083E-01	2403.0	0.8688E+05	1104.0	107.0	795.0	1602.0	0.1089E+05	795.822	1104.0	54.2967
28	0.8480E-01	2023.0	0.9936E+05	1088.0	106.0	720.0	1235.0	0.1134E+05	721.152	1088.0	76.0979
29	0.1110	985.0	0.9662E+05	1052.0	77.4	754.0	1636.0	0.1093E+05	755.434	1052.0	76.6925
30	0.1180	1720.0	0.1149E+06	1012.0	76.3	716.0	1406.0	0.1161E+05	718.123	1012.0	106.151
31	0.8566E-01	4939.0	0.9829E+05	995.0	69.8	742.0	1298.0	0.1036E+05	743.158	994.999	46.3359
32	0.1050	1293.0	0.8778E+05	1101.0	108.0	797.0	1478.0	0.1032E+05	798.017	1101.0	73.2740
33	0.8969E-01	3494.0	0.1139E+06	1042.0	101.0	790.0	1601.0	0.1061E+05	790.701	1042.0	42.0874
34	0.1090	1894.0	0.6707E+05	1061.0	84.4	789.0	1437.0	0.1126E+05	790.457	1061.0	79.1258
35	0.1250	4345.0	0.6927E+05	1021.0	78.6	762.0	1669.0	0.1135E+05	763.818	1021.0	85.8513
36	0.9419E-01	4163.0	0.9238E+05	1107.0	66.7	804.0	1123.0	0.1049E+05	806.000	1107.0	78.3645
37	0.9908E-01	779.0	0.1102E+06	1084.0	65.3	758.0	1569.0	0.1000E+05	759.395	1084.0	63.8048
38	0.1150	5981.0	0.7600E+05	1077.0	92.9	807.0	1151.0	0.1029E+05	808.853	1077.0	99.5990
39	0.1270	1220.0	0.7380E+05	1049.0	91.6	769.0	1207.0	0.1007E+05	770.873	1049.0	117.576
40	0.6327E-01	3603.0	0.7221E+05	1032.0	73.0	740.0	1459.0	0.9940E+04	740.596	1032.0	24.9673
41	0.8654E-01	1035.0	0.8923E+05	1000.0	99.0	774.0	1336.0	0.1097E+05	774.657	999.999	43.5338
42	0.8840E-01	1412.0	0.8486E+05	1083.0	111.0	775.0	1171.0	0.1152E+05	776.029	1083.0	74.1388
43	0.1380	2896.0	0.8553E+05	1028.0	79.1	731.0	1267.0	0.1041E+05	733.895	1028.0	144.571
44	0.9766E-01	3368.0	0.7792E+05	1075.0	97.0	709.0	1632.0	0.1057E+05	710.214	1075.0	70.7905
45	0.1070	910.0	0.8256E+05	1067.0	71.9	765.0	1377.0	0.1192E+05	766.872	1067.0	93.4468
46	0.9668E-01	5456.0	0.6340E+05	1066.0	98.4	718.0	1191.0	0.1190E+05	719.797	1066.0	101.575
47	0.1070	1569.0	0.1083E+06	1096.0	67.7	771.0	1328.0	0.1161E+05	773.289	1096.0	101.476
48	0.8296E-01	8875.0	0.6732E+05	1056.0	85.6	757.0	1420.0	0.1024E+05	758.000	1056.0	46.4636
49	0.1020	2523.0	0.7028E+05	1031.0	115.0	763.0	1365.0	0.1108E+05	763.992	1031.0	70.8401
50	0.7149E-01	605.0	0.1109E+06	1093.0	93.6	706.0	1562.0	0.1044E+05	706.638	1093.0	41.4624

APPENDIX B - STEPWISE REGRESSION

Stepwise regression is one of several statistical techniques commonly employed to identify the "best" subset of parameters that can be used to predict a performance measure of interest (e.g., Q). The procedure for this technique may be summarized as follows:

- 1. Select the variable (parameter by itself, crossproduct, or higher order term depending on assumed model for Q) with the largest absolute linear correlation with Q.
- 2. Using partial correlation (instead of the simple linear correlation in step 1), select the variable with the highest absolute partial correlation. This partial correlation with Q accounts for the variables already entered into the stepwise model being formed. If this variable's partial correlation meets a specified entry significance level, enter it into the model; otherwise, terminate the stepwise regression.
- 3. Given the variable in step 2 that just entered the regression equation, calculate the partial correlation of all variables previously entered into the model. If any of them no longer meet a specified removal significance level, delete them from the regression equation.
- 4. Go to step 2.

APPENDIX C - DETERMINISTIC SENSITIVITY COEFFICIENTS FOR THE N = 10 DESIGN MATRIX

This appendix presents the detailed results of the deterministic sensitivity analysis based on the N = 10 LHS generaged design matrix (Table C-1). For each of the 10 design points, the three dependent variables (Q, H_{wl} , and H_{wu}) are listed at the top. The eight input parameters follow with their values (generated by LHS), first partial derivative sensitivity coefficient $(\partial R/\partial \alpha)$, and normalized sensitivity coefficient $[(\partial R/\partial \alpha)(\alpha/R)]$.

Table C-1. Detailed Results of the Deterministic Sensitivity Analysis.

Parameter	Value	<u>∂R</u> ∂α	<u> </u>	
1	R=Q = 48.52464	H _{W]} = 783.7536	H _{wu} = 1043.999	
Tu	83370.00	0.2156935E-08	0.3705821E-05	
Н <mark>и</mark>	1044.00	0.1859182	4.000000	
r	2948.00	-0.4557747E-05	-0.2758951E-03	
	0.08609	1124.082	1.994291	
rw Tl Hl	107.00	0.1309451E-02	0.2887424E-02	
Η'n	783.00	-0.1859182	-3.000000	
Κw	10010.00	0.4833502E-02	0.9971089	
٤"	1250.00	-0.3870748E-01	-0.9971089	
2	R=Q = 99.88144	$H_{w]} = 790.5381$	$H_{wu} = 1092.998$	
Tu	88400.00	0.7200674E-08	0.6372952E-05	
หน	1093.00	0.3274801	3.583607	
r	5194.00	-0.1481603E-04	-0.7704580E-03	
	0.105	1886.844	1 .9 83538	
r _w Tj Hj	67.70	0.1227723E-01	0.8321550E-02	
нί	788.00	-0.3274801	-2.583607	
Kw	11260.00	0.8796593E-02	0.9916721	
L"	1181.00	-0.8386930E-01	-0.9916721	

Table C-1. (Continued).

Parameter	Value	<u>∂R</u> ∂α	aR α aα R
3	R=Q = 81.88855	H _{w1} = 759.2360	H _{wu} = 992,9987
Tu	94710.00	0.4734320E-08	0.5475582E-05
Hu	993.00	0.3484619	4.225532
r	1358.00	-0.3395267E-04	-0.5630546E-03
rw	0.118	1380.910	1.989868
Tl	98.60	0.4368129E-02	0.5259557E-02
ΗJ	758.00	-0.3484619 0.6938450E-02	-3.225532 0.9947350
K _W L	11740.00 1466.00	-0.5556440E-01	-0.9947350
4	R=Q = 39.77062	$H_{w1} = 811.6145$	$H_{wu} = 1037.00$
Tu	109100.00	0.7377022E-09	0.2023687E-0
Н _и	1037.00	0.1759762	4.588496
٣	240.00	-0.5719953E-04 877.3659	-0.3451766E-0 1.996489
Ϋ́W T-	0.0905 81.20	0.1331737E-02	0.2719018E-0
Т _Т Н 7	811.00	-0.1759762	-3.588496
Κ̈́w	10520.00	0.3770191E-02	0.9972790
L	1534.00	-0.2585555E-01	-0.9972790
5	R=Q = 71.92241	$H_{W1} = 712.1008$	H _{wu} = 1100.99
Tu	92190.00	0.2460308E-08	0.3153618E-0
Hu	1101.00	0.1844164	2.823077
٣	1861.00	-0.1102517E-04	-0.2852775E-0
r _w	0.09287	1544.613 0.1970982E-02	1.994485 0.2822641E-0
T p H p	103.00 711.00	-0.1844164	-1.823077
K _W	10750.00	0.6671550E-02	0.9971741
L	1575.00	-0.4553598E-01	-0.9971741
6	R=Q = 120.8852	$H_{W1} = 736.0329$	$H_{WU} = 1054.9$
Tu	101500.00	0.6427935E-08	0.5397149E-0
Hů	1055.00	0.3765893	3.286604
۳	1165.00	-0.7195967E-04	-0.6934929E-0
Ϋ́Ψ T_	0.125 86.50	1922.578 0.8850572E-02	1.988021 0.6333071E-0
Т _l Нj	734.00	-0.3765893	-2.286604
Kw	10230.00	0.1174183E-01	0.9936614
L	1325.00	-0.9065580E-01	-0.9936614

Table C-1. (Continued).

		·· (continued).	
Paramete	r Value	∂R ∂α	<u> </u>
7	R=Q = 86.32016	$H_{w1} = 716.6310$	H _{wu} = 1071.998
T _u Hu	67980.00 1072.00	0.6562483E-08 0.2417931	0.5168174E-05 3.002801
r r _w Ti	902.00 0.09778 76.90	-0.4794403E-04 1757.957 0.5128352E-02	-0.5009897E-03 1.991343 0.4568693E-02
Tī Hī Kw	715.00 11160.00	-0.2417931 0.7699403E-02	-2.002801 0.9954261
L	1380.00	-0.6226474E-01	-0.9954261
8	R=Q = 86.88659	$H_{w1} = 800.2331$	$H_{wu} = 1084.998$
T _u	69440.00 1085.00 2616.00	0.8779252E-08 0.3037992 -0.1423474E-04	0.7016403E-05 3.793706 -0.4285827E-03
r r _w Ti	0.11 113.00	1573.013 0.3315279E-02	1.991463 0.4311673E-02
Tj Hj K _w	799.00 11470.00	-0.3037993 0.7542402E-02	-2.793706 0.9956813
L	1429.00	-0.6053978E-01	-0.9956813
9	R=Q = 46.98318	$H_{w1} = 769.2566$	$H_{wu} = 1008.999$
T _u H _u	74270.00 1009.00	0.31442505E-08 0.1949509	0.4970363E-05 4.186722
r r _w Ta	14690.00 0.10 70.80	-0.1402964E-05 934.7742 0.3460010E-02	-0.4386581E-03 1.989593 0.5213967E-02
T _] H _] K _w	768.00 10430.00	-0.1949509 0.4481110E-02	-3.186722 0.9947811
Ľ	1672.00	-0.2795333E-02	-0.9947811
10	•	$H_{w1} = 739.0709$	$H_{wu} = 1024.999$
T _u H _u	115300.00 1025.00	0.1338124E-08 0.1833462	0.2932052E-05 3.571428
r Yw	8017.00 0.07457	-0.2115667E-05 1406.067	-0.3223333E-03 1.992582
Tĵ H1	90.60 738.00	0.2167197E-02 -0.1833462	0.3731408E-02 -2.571428
K W L	12020.00 1141.00	0.4361387E-02 -0.4594555E-01	0.9962656 -0.9962656

APPENDIX D - RANGE OF DETERMINISTIC NORMALIZED SENSITIVITY COEFFICIENTS

Table D-1 gives the deterministic normalized sensitivity coefficients for the extremes of the range of the independent parameters as shown in Table 2-1.

Table D-1. Normalized Sensitivity Coefficients for the Performance Measure Q.

Input Parameter	Minimum of Range	Maximum of Range
r _w	1.999	1.982
r	-0.174 x 10 ⁻⁷	-0.690×10^{-3}
Tu	0.213 x 10 ⁻⁹	0.880×10^{-5}
Hu-H1	1.0	1.0
т	0.213 x 10 ⁻³	0.877×10^{-2}
L	-1.0	-0.991
K _W	1.0	0.991

For the key parameters $(r_W, H_U-H_I, L, and K_W)$, it is seen that these coefficients differ little at the extremes of the ranges studied. The other parameters remain unimportant over their respective ranges.

DISTRIBUTION LIST

ACRES AMERICAN INC

A. S. BURGESS

ROBERT H. CURTIS

AEROSPACE CORP

BARRETT R. FRITZ

AGBABIAN ASSOCIATES

CHRISTOPHER M. ST JOHN

ALLIED GENERAL NUCLEAR SERVICES

K. J. ANDERSON

AMARILLO PUBLIC LIBRARY

AMERICAN EMBASSY - SWEDEN

ANALYSIS AND TECHNOLOGY INC

T. MAZOUR

APPLIED MECHANICS INC

GRAHAM G. MUSTOE

JOHN R. WILLIAMS

ARGONNE NATIONAL LABORATORY

DAVID F. FENSTER

WYMAN HARRISON

I. HOWARD KITTEL

MARTIN SEITZ

MARTIN J. STEINDLER

ARINC RESEARCH CORP

H. P. HIMPLER

ARIZONA PUBLIC SERVICE COMPANY

HENRY W. RILEY, JR.

ARTHUR D. LITTLE INC

AVIVA BRECHER

CHARLES R. HADLOCK

ATKINS RESEARCH & DEVELOPMENT - UNITED

KINGDOM

T. W. BROYD

ATOMIC ENERGY CONSULTANTS

DONALD G. ANDERSON

ATOMIC ENERGY CONTROL BOARD -

CANADA

KEN SHULTZ

ATOMIC ENERGY OF CANADA LTD

T. CHAN

ANN QUINN

F. P. SARGENT

ATOMIC ENERGY RESEARCH ESTABLISHMENT

- UNITED KINGDOM

D. P. HODGKINSON

AUSTRALIAN ATOMIC ENERGY COMMISSION

BABCOCK & WILCOX

INFORMATION SERVICES

BATTELLE COLUMBUS DIVISION

SANFORD G. BLOOM

JOHN T. MCGINNIS

JEFFREY L. MEANS

NEIL E. MILLER

STEPHEN NICOLOSI

RALPH THOMAS

KENNETH R. YATES

BECHTEL GROUP INC

LESLIE 1. JARDINE

N. A. NORMAN

BENDIX FIELD ENGINEERING CORP

MICHAEL H. MOBLEY

JOHN C. PACER

BERKELEY GEOSCIENCES/HYDROTECHNIQUE

ASSOCIATES

BRIAN KANEHIRO

BLACK & VEATCH

M. JOHN ROBINSON

BOEING ENGINEERING AND CONSTRUCTION

COMPANY

R. B. CAIRNS

BRENK SYSTEMPLANUNG - W. GERMANY

H. D. BRENK

BRIGHAM YOUNG UNIVERSITY

HAROLD B. LEE LIBRARY

BROOKHAVEN NATIONAL LABORATORY

M. S. DAVIS

CLAUDIO PESCATORE

PETER SOO

BUNDESANSTALT FUR GEOWISSENSCHAFTEN

UND ROHSTOFFE - W. GERMANY

MICHAEL LANGER

BUNDESMINISTERIUM FUR FORSCHUNG UND

TECHNOLOGIE - W. GERMANY

ROLF-PETER RANDL

BUREAU DE RECHERCHES GEOLOGIQUES ET

MINIERES - FRANCE

PIERRE F. PEAUDECERF

CALIFORNIA ASSEMBLY COMMITTEE ON

NATURAL RESOURCES

GENE VARANINI

CAPITAL AREA GROUND WATER

CONSERVATION COMMISSION

A. N. TURCAN, JR.

CENTRE D INFORMATIQUE GEOLOGIQUE -

FRANCE

GHISLAIN DEMARSILY

CHALMERS UNIVERSITY OF TECHNOLOGY -

SWEDEN

BERT ALLARD

CITIZENS INSTITUTE FOR A POSITIVE ENERGY

POLICY

LINDSAY AUDIN

COLORADO OUTWARD BOUND SCHOOL

DAVID L. BURGER

PETER ANTHONY ONEIL

COLORADO SCHOOL OF MINES

W. HUSTRULID

CONGRESSIONAL INFORMATION SERVICE

LINDLEY C. MCGREW

CORTLAND COUNTY HEALTH DEPT

DAPPOLONIA CONSULTING ENGINEERS INC LISA K. DONOHUE

ABBY FORREST

AMINA HAMDY

CARL E. SCHUBERT DAWCON MANAGEMENT CONSULTING

DAVID A. WEBSTER

DEAF SMITH COUNTY LIBRARY

DEPT OF ENERGY, MINES AND RESOURCES -

CANADA

A. S. JUDGE

DYNATECH R/D COMPANY

STEPHEN E. SMITH

E.I. DU PONT NEMOURS & COMPANY

D. H. TURNO

E.L.H. PUBLICATIONS - THE RADIOACTIVE

EXCHANGE

HELMINSKI & WILKEN E.R. IOHNSON ASSOCIATES INC

E. R. JOHNSON

G. L. JOHNSON

EAST COMPANY INC

RAYMOND PEREZ

EBASCO SERVICES INC **ZUBAIR SALEEM**

RAYMOND H. SHUM **ECOLOGY CENTER OF LOUISIANA**

ROSS VINCENT

EDISON ELECTRIC INSTITUTE

R. E. L. STANFORD

EDS NUCLEAR INC

C. SUNDARARAJAN

EG & G IDAHO INC

ROGER A. MAYES

M. D. MCCORMACK ROBERT M. NEILSON, JR.

ELECTRIC POWER RESEARCH INSTITUTE

CHAIM BRAUN

ELSAM - DENMARK

A, V. JOSHI

ARNE PEDERSEN

ENERGY RESEARCH GROUP INC

MARC GOLDSMITH

ENGINEERS INTERNATIONAL INC

FRANCIS S. KENDORSKI

ENVIRONMENTAL POLICY INSTITUTE

DAVID M. BERICK

ENVIROSPHERE COMPANY

ROGER G. ANDERSON

K. E. LIND-HOWE

EXXON NUCLEAR COMPANY INC GERALD L. RITTER

EXXON NUCLEAR IDAHO COMPANY INC

NATHAN A. CHIPMAN GARY WAYMIRE

FENIX & SCISSON INC

JOSE A. MACHADO

CHARLENE U. SPARKMAN

FLORIDA INSTITUTE OF TECHNOLOGY JOSEPH A. ANGELO, JR.

FLORIDA POWER & LIGHT COMPANY JAMES R. TOMONTO

FONTANA CORROSION CENTER

DIGBY MACDONALD

FORD, BACON & DAVIS INC ROBERT D. BAIRD

DARRELL H. CARD

ROBERT F. OVERMYER BURTON J. THAMER

FOSTER-MILLER ASSOCIATES INC

NORBERT PAAS GENERAL ATOMIC COMPANY

MICHAEL STAMATELATOS

GENERAL COURT OF MASSACHUSETTS TIMOTHY J. BURKE

GEOLOGICAL SURVEY OF CANADA

JEFFREY HUME

GEORESULTS INC DAVID SNOW

GEORGIA INSTITUTE OF TECHNOLOGY

GEOFFREY G. EICHHOLZ ALERED SCHNEIDER

GEOTHERMAL ENERGY INSTITUTE

DONALD F. X. FINN **GEOTRANS**

IAMES MERCER

GESELLSCHAFT F. STRAHLEN U. **UMWELTFORSCHUNG M.B.H. - W.**

GERMANY

WOLFGANG BODE H. MOSER

GILBERT/COMMONWEALTH JERRY L. ELLIS

GOLDER ASSOCIATES DENNIS GOLDMAN

MELISSA MATSON

I. W. VOSS GRAND COUNTY PUBLIC LIBRARY

GTC GEOLOGIC TESTING CONSULTANTS LTD -

CANADA

JOHN F. PICKENS GULF INTERSTATE INC

THOMAS J. HILL

GULF STATES UTILITIES COMPANY KERNFORSCHUNGSZENTRUM KARLSRUHE MICHAEL J. GILBERT GMBH - W. GERMANY E. LINN DRAPER OSWALD H. GREAGER H & R TECHNICAL ASSOCIATES INC K. D. CLOSS KENNETH GUSCOTT WILLIAM R. RHYNE R. KOESTER ROBERT HIGGINS HAHN-MEITNER-INSTITUT FUR KIHN ASSOCIATES CRAIG W. JONES KERNFORSCHUNG BERLIN HARRY KIHN TERRY R. LASH KLAUS ECKART MAASS KLM ENGINEERING INC MAX MCDOWELL HANFORD ENGINEERING DEVELOPMENT **B. GEORGE KNIAZEWYCZ** A. ALAN MOGHISSI KOREA INSTITUTE OF ENERGY AND LABORATORY BARBARA MORRA ROBERT EINZIGER RESOURCES (KIER) CAROLINE PETTI W. E. ROAKE CHONG SU KIM MARTIN RATHKE HATTIESBURG PUBLIC LIBRARY **KYOTO UNIVERSITY - JAPAN** TOM & MARY REES HIGH PLAINS WATER DISTRICT YORITERU INOUE TIM REVELL DON MCREYNOLDS LAKE SUPERIOR REGION RADIOACTIVE OWEN SEVERANCE HOUGH-NORWOOD HEALTH CARE CENTER **WASTE PROJECT** PATRICIA SNYDER GEORGE H. BROWN, M.D. C. DIXON P. E. STRALEY-GREGA **ILLINOIS STATE GEOLOGICAL SURVEY** LAWRENCE BERKELEY LABORATORY MARGUERITE SWEENEY ROBERT E. BERGSTROM **IOHN A. APPS** M. J. SZULINSKI KEROS CARTWRIGHT **EUGENE BINNALL** MARTIN & ELAINE WALTER IMPERIAL COLLEGE OF SCIENCE AND I. WANG LINDA WITTKOPF **TECHNOLOGY - ENGLAND** LAWRENCE LIVERMORE NATIONAL STEPHEN G. ZEMBA B. K. ATKINSON LABORATORY MICHAEL BAKER, JR. INC **INSTITUT FUR TIEFLAGERUNG - W. GERMANY** THOMAS E. MCKONE C. J. TOUHILL WERNT BREWITZ LAWRENCE D. RAMSPOTT (2) MICHIGAN DEPT OF NATURAL RESOURCES KLAUS KUHN W. G. SUTCLIFFE R. THOMAS SEGALL **INTER/FACE ASSOCIATES INC** TECHNICAL INFORMATION DEPARTMENT MICHIGAN DEPT OF PUBLIC HEALTH RON GINGERICH L-53 GEORGE W. BRUCHMANN INTERA ENVIRONMENTAL CONSULTANTS INC LOS ALAMOS NATIONAL LABORATORY LEE E. JAGER MICHIGAN ENVIRONMENTAL PROTECTION F. J. PEARSON, JR. P. L. BUSSOLINI LARRY RICKERTSEN WAYNER, HANSEN COMMITTEE ROBERT WILEMS W. C. MYERS DAVE CHAPMAN INTERNATIONAL ATOMIC ENERGY AGENCY -K. K. S. PILLAY MICHIGAN PUBLIC SERVICE COMMISSION LOS ALAMOS TECHNICAL ASSOCIATES INC RON CALLEN FRANK A. OHARA R. J. KINGSBURY MINNESOTA ENVIRONMENTAL QUALITY INTERNATIONAL ENERGY ASSOCIATES LTD LOUISIANA DEPT OF TRANSPORTATION & BOARD BLYTHE K. LYONS DEVELOPMENT RICHARD PATON INTERNATIONAL ENERGY SYSTEMS CORP GEORGE H. CRAMER, II MINNESOTA GEOLOGICAL SURVEY JOHN A. BOWLES **LOUISIANA TECHNICAL UNIVERSITY** LINDA L. LEHMAN INTERNATIONAL ENGINEERING COMPANY R. H. THOMPSON MATT S. WALTON **LUBBOCK COUNTY SOIL AND WATER** INC MINNESOTA STATE ENERGY AGENCY TERRY L. STEINBORN CONSERVATION DISTRICT MISSISSIPPI ATTORNEY GENERALS OFFICE MAX ZASLAWSKY **DON LANGSTON** MACK CAMERON INTERNATIONAL RESEARCH AND MAINE BUREAU OF HEALTH MISSISSIPPI BUREAU OF GEOLOGY **EVALUATION** DONALD C. HOXIE MICHAEL B. E. BOGRAD MARYLAND DEPT OF HEALTH & MENTAL R. DANFORD MISSISSIPPI CITIZENS AGAINST NUCLEAR **IOWA STATE UNIVERSITY** DISPOSAL MARTIN C. EDELSON MAX FISENBERG STANLEY DEAN FLINT **BERNARD I. SPINRAD** MASSACHUSETTS HOUSE OF MISSISSIPPI DEPT OF ENERGY AND IRT CORP REPRESENTATIVES TRANSPORTATION J. STOKES WILLIAM ROBINSON **RONALD J. FORSYTHE** ISTITUTO SPERIMENTALE MODELLI E MISSISSIPPI DEPT OF NATURAL RESOURCES MASSACHUSETTS INSTITUTE OF STRUTTURE S.P.A. - ITALY **TECHNOLOGY** CHARLES L. BLALOCK **IOHN DEUTCH** CURTIS W. STOVER JACKSON STATE UNIVERSITY **TED GREENWOOD** MISSISSIPPI EMERGENCY MANAGEMENT **ESTUS SMITH** MARSHA LEVINE AGENCY JACKSON-GEORGE REGIONAL LIBRARY MCDERMOTT INTERNATIONAL JAMES E. MAHER JAY L. SMITH COMPANY INC. KAREN L. FURLOW MISSISSIPPI STATE BOARD OF HEALTH JAY L. SMITH MCMASTER UNIVERSITY - CANADA **EDDIE S. FUENTE** JGC CORPORATION - JAPAN L. W. SHEMILT **GUY R. WILSON** MASAHIKO MAKINO MEMBERS OF THE GENERAL PUBLIC MISSISSIPPI STATE HOUSE OF JOHNS HOPKINS UNIVERSITY L. ROBERT ANDERSON REPRESENTATIVES IARED L. COHON KURT BALLING HILLMAN TEROME ERAZIER KAISER ENGINEERS INC THOMAS G. BRADFORD MISSISSIPPI STATE UNIVERSITY W. J. DODSON ROGER H. BROOKS VICTOR L. ZITTA KALAMAZOO COLLEGE HAZEL CHAPMAN, PH.D. MITRE CORP RALPH M. DEAL LAWRENCE CHASE, PH.D. LESTER A. ETTLINGER KANSAS DEPT OF HEALTH AND TOM & SUSAN CLAWSON MITSUBISHI METAL CORP **ENVIRONMENT** STEVE CONEWAY TATSUO ARIMA GERALD W. ALLEN M. VAL DALTON MOAB NUCLEAR WASTE INFORMATION KARNBRANSLESAKERHET - SWEDEN **KENNETH & ALICE M. DROGIN**

DANNELLE D. DUDEK

T. W. EDWARDS, JR.

THAUMAS P. EHR BOB GAMMELIN MICHAELENE PENDLETON (2)

LARS B. NILSSON

FRANK WREATH

KELLER WREATH ASSOCIATES

RONALD C. ARNETT CATHY S. FORE **MONTANA BUREAU OF MINES AND** HARRY BABAD GEOLOGY DAVID C. KOCHER KARL M. LA RUE F. M. OBLOW **EDWARD C. BINGLER** MICHAEL I. SMITH MONTICELLO NUCLEAR WASTE **ELLEN D. SMITH** RICHARD T. WILDE **OHIO STATE UNIVERSITY INFORMATION OFFICE ROCKWELL INTERNATIONAL ENERGY SYSTEMS** J. B. NEUHARDT CARL EISEMANN (2) GROUP NAGRA - SWITZERLAND M. A. FLIGNER ONTARIO HYDRO - CANADA LAWRENCE J. SMITH HANS ISSLER ROGERS & ASSOCIATES ENGINEERING CORP **NATIONAL ACADEMY OF SCIENCES** J. A. CHADHA ARTHUR SUTHERLAND JOHN T. HOLLOWAY K. A. CORNELL ROSS LABS NATIONAL AERONAUTICS AND SPACE C. F. LEE ONTARIO MINISTRY OF THE ENVIRONMENT -**CHARLES PAULE ADMINISTRATION** W. T. MALONE MICHAEL R. HELFERT CANADA LARRY LUTZ JAAK VIIRLAND MICHAEL ZOLENSKY OREGON STATE UNIVERSITY ROY F. WESTON INC NATIONAL HYDROLOGY RESEARCH MARTIN HANSON **JOHN C. RINGLE** INSTITUTE - CANADA ORGANISATION FOR ECONOMIC RONALD MACDONALD **DENNIS J. BOTTOMLEY COOPERATION AND DEVELOPMENT -**MICHAEL V. MELLINGER **NATIONAL PARKS & CONSERVATION** ROBERT SCHULER **ASSOCIATION** FRANCE PETER D. JOHNSTON HARRY W. SMEDES T. DESTRY JARVIS **ROYAL INSTITUTE OF TECHNOLOGY -**OTHA INC TERRI MARTIN **SWEDEN** NATIONAL SCIENCE FOUNDATION JOSEPH A. LIEBERMAN **ROGER THUNVIK** PACIFIC NORTHWEST LABORATORY ROYAL E. ROSTENBACH **NAVAL WEAPONS STATION EARLE** DON J. BRADLEY RPC INC JAMES VANCE H. C. BURKHOLDER **GENNARO MELLIS** S.E. LOGAN & ASSOCIATES INC **NEW MEXICO ENVIRONMENTAL EVALUATION** JOHN B. BURNHAM JOE DEVARY STANLEY E. LOGAN GROUP S.M. STOLLER CORP ROBERT H. NEILL HARVEY DOVE ROBERT W. KUPP **NEW YORK DEPT OF HEALTH** FLOYD N. HODGES SALT LAKE CITY TRIBUNE DAVID AXELROD, M.D. J. H. JARRETT IIM WOOLF **NEW YORK ENERGY RESEARCH &** MAX R. KREITER SAN DIEGO GAS & ELECTRIC COMPANY I. E. MENDEL **DEVELOPMENT AUTHORITY** JOHN P. SPATH (8) R. E. NIGHTINGALE LOUIS BERNATH SAN JUAN RECORD **NEW YORK GEOLOGICAL SURVEY** J. M. RUSIN **DEBORAH A. MARCUS ROBERT H. FAKUNDINY** R. JEFF SERNE PARSONS BRINCKERHOFF QUADE & SANDIA NATIONAL LABORATORIES **NEW YORK POWER AUTHORITY** MARGARET S. CHU MYRON M. KACZMARSKY **DOUGLAS INC** NANCY C. FINLEY **NEW YORK STATE GEOLOGICAL SURVEY** T. R. KUESEL **RON IMAN** JAMES R. ALBANESE ROBERT PRIETO J. KEITH JOHNSTONE **NEW YORK STATE HEALTH DEPT** MARK E. STEINER R. W. LYNCH PENNSYLVANIA STATE UNIVERSITY JOHN MATUSZEK MARTIN A. MOLECKE **NEW YORK STATE PUBLIC SERVICE** WILLIAM B. WHITE JAMES T. NEAL PERRY COUNTY CITIZENS AGAINST NUCLEAR COMMISSION **NESTOR R. ORTIZ FRED HAAG** WASTE DISPOSAL NORTHEAST UTILITIES SERVICE COMPANY SCOTT SINNOCK WARREN STRICKLAND LYNN D. TYLER PATRICIA ANN OCONNELL PETTIS WALLEY WENDELL D. WEART **NORTHWESTERN UNIVERSITY** PHYSIKALISCH-TECHNISCHE BUNDESANSTALT WIPP CENTRAL FILES **BERNARD J. WOOD** W. GERMANY **SARGENT & LUNDY ENGINEERS NTR GOVERNMENT SERVICES** PETER BRENNECKE LAWRENCE L. HOLISH THOMAS V. REYNOLDS PIRGIM SAVANNAH RIVER LABORATORY **NUCLEAR ASSURANCE CORP** RICHARD LEVICK CAROL JANTZEN JOHN V. HOUSTON POINT BEACH NUCLEAR PLANT WILLIAM R. MCDONELL JEAN RION JAMES J. ZACH DONALD ORTH **NUCLEAR ENERGY AGENCY/OECD - FRANCE PUBLIC SERVICE INDIANA** SCANDPOWER INC **ANTHONY MULLER ROBERT S. WEGENG** DAN POMEROY **NUCLEAR WASTE WATCHERS** PURDUE UNIVERSITY SCIAKY BROTHERS **HELEN LETARTE** PAUL S. LYKOUDIS JOHN C. JASPER RALPH M. PARSONS COMPANY **NUS CORP** SCIENCE APPLICATIONS INC W. G. BELTER JERROLD A. HAGEL JEFFREY ARBITAL **RODNEY J. DAVIS** RE/SPEC INC JERRY J. COHEN N. BARRIE MCLEOD GARY D. CALLAHAN NADIA DAYEM WILLIAM C. MCCLAIN DOUGLAS D. ORVIS **BARRY DIAL** YONG M. PARK **RED ROCK 4-WHEELERS** JAMES E. HAMMELMAN GEORGE SCHULTZ DOUGLAS W. TONKAY DEAN C. KAUL **NUTECH ENGINEERS INC** RENSSELAER POLYTECHNIC INSTITUTE J. ROBERT LARIVIERE **GARRISON KOST** MARTY BECKER DAVID H. LESTER **NWT CORP IAMES WU** PETER E. MCGRATH **RESOURCE SYSTEMS INSTITUTE** W. L. PEARL JOHN E. MOSIER OAK RIDGE NATIONAL LABORATORY KIRK R. SMITH HOWARD PRATT RHODE ISLAND GOVERNORS OFFICE J. O. BLOMEKE MICHAEL E. SPAETH **IOHN A. IVEY** DAN CACUCI M. D. VOEGELE **RIO ALGOM CORP** H. C. CLAIBORNE KRISHAN K. WAHI DUANE MATLOCK ALLEN G. CROFF

ROCKWELL HANFORD OPERATIONS

LESLIE R. DOLE

ROBERT A. YODER

DEVELOPMENT

SENECA COUNTY DEPT OF PLANNING &

SHANNON & WILSON INC

HARVEY W. PARKER

SHIMIZU CONSTRUCTION COMPANY LTD

IUNII TAKAGI

SHIMIZU CONSTRUCTION COMPANY LTD -

JAPAN

TAKASHI ISHII

SIERRA CLUB

MARVIN RESNIKOFF

SIERRA CLUB - COLORADO OPEN SPACE

COUNCIL

ROY YOUNG

SLICKROCK COUNTRY COUNCIL

BRUCE HUCKO

SNAKE RIVER ALLIANCE

TIM MCNEIL

SOGO TECHNOLOGY INC

TIO C. CHEN

SOUTH DAKOTA GEOLOGICAL SURVEY

RICHARD BRFTZ

SOUTHERN STATES ENERGY BOARD

J. F. CLARK

SOUTHWEST RESEARCH AND INFORMATION

CENTER

DON HANCOCK ALISON P. MONROE

SPRINGVILLE CITY LIBRARY

STANFORD UNIVERSITY

GEORGE A. PARKS

IRWIN REMSON

STONE & WEBSTER ENGINEERING CORP

SUE NEWHAMS

JOHN H. PECK

EVERETT M. WASHER

STUDSVIK ENERGITEKNIK AB - SWEDEN

ROLF SJOBLOM

SWANSON ENVIRONMENTAL INC

PETER G. COLLINS

SYRACUSE UNIVERSITY

WALTER MEYER

SYSTEM DEVELOPMENT CORP

RHONNIE L. SMITH

SYSTEMS SCIENCE AND SOFTWARE

PETER LAGUS

TECHNICAL INFORMATION PROJECT

DONALD PAY

TECHNICAL RESEARCH CENTRE OF FINLAND

SEPPO VUORI

TEKNEKRON RESEARCH INC

DOUGLAS K. VOGT

TEXAS A & M UNIVERSITY STEVE MURDOCK

JAMES E. RUSSELL

TEXAS DEPT OF HEALTH

DAVID K. LACKER

TEXAS DEPT OF WATER RESOURCES

C. R. BASKIN

TEXAS GOVERNORS OFFICE OF GENERAL

COUNSEL

R. DANIEL SMITH

THE ANALYTIC SCIENCES CORP

IOHN W. BARTLETT

CHARLES M. KOPLIK

THE EARTH TECHNOLOGY CORP

JOSEPH G. GIBSON

MATT WERNER

THE JACKSON CLARION-LEDGER

MARK SCHIFFSTFIN

TRU WASTE SYSTEMS OFFICE

K. V. GILBERT

U.H.D.E. - W. GERMANY

FRANK STEINBRUNN

U.S. BUREAU OF LAND MANAGEMENT

MARY PLUMB

U.S. BUREAU OF RECLAMATION

JOHN BROWN

REGE LEACH

U.S. DEPT OF ENERGY

CHED BRADLEY

R. COOPERSTEIN

LAWRENCE H. HARMON

CARL NEWTON

IAMES TURI

U.S. DEPT OF ENERGY - ALBUQUERQUE

OPERATIONS OFFICE

PHILIP LARRAGOITE

JOSEPH M. MCGOUGH

U.S. DEPT OF ENERGY - CHICAGO

OPERATIONS OFFICE

NURI BULUT

GARY C. MARSHALL

PUBLIC READING ROOM

R. SELBY

U.S. DEPT OF ENERGY - GEOLOGIC

REPOSITORY DIVISION

C. R. COOLEY (2)

WARREN EISTER

J. FIORE

MARK W. FREI

RALPH STEIN

U.S. DEPT OF ENERGY - GRAND JUNCTION

OFFICE

WAYNE ROBERTS

U.S. DEPT OF ENERGY - HEADQUARTERS

PUBLIC READING ROOM

HENRY F. WALTER

U.S. DEPT OF ENERGY - IDAHO OPERATIONS

OFFICE

M. BARA: NCA

JAMES F. .EONARD

PUBLIC READING ROOM U.S. DEPT OF THERGY - NEVADA OPERATIONS

PUBLIC F. - DING ROOM

U.S. DEPT (18) HERGY - NWTS PROGRAM

OFFICE

J. O. NEFF

U.S. DEPT OF ENERGY - OAK RIDGE

OPERATIONS OFFICE

PUBLIC READING ROOM

U.S. DEPT OF ENERGY - RICHLAND

OPERATIONS OFFICE

J. SCHREIBER

U.S. DEPT OF ENERGY - SAN FRANCISCO

OPERATIONS OFFICE

ENERGY RESOURCES CENTER PUBLIC READING ROOM

U.S. DEPT OF ENERGY - TECHNICAL

INFORMATION CENTER (317)

U.S. ENVIRONMENTAL PROTECTION AGENCY

DIVISION OF CRITERIA & STANDARDS IAMES NEIHEISEL

U.S. GENERAL ACCOUNTING OFFICE WILLIAM DAVID BROOKS

U.S. GEOLOGICAL SURVEY - ALEXANDRIA

G. N. RYALS U.S. GEOLOGICAL SURVEY - COLUMBUS

A. M. LA SALA, JR.

U.S. GEOLOGICAL SURVEY - JACKSON

GARALD G. PARKER, JR.

U.S. GEOLOGICAL SURVEY - RESTON

JOHN ROBERTSON

EUGENE H. ROSEBOOM, JR.

DAVID B. STEWART

NEWELL J. TRASK, JR.

U.S. HOUSE SUBCOMMITTEE ON ENERGY AND

THE ENVIRONMENT

MORRIS K. UDALL

U.S. NUCLEAR REGULATORY COMMISSION

I. CALVIN BELOTE

EILEEN CHEN

ENRICO F. CONTI

F. R. COOK

DOCKET CONTROL CENTER

PAUL F. GOLDBERG

PHILIP 5. JUSTUS

MALCOLM R. KNAPP

JOHN C. MCKINLEY

THOMAS J. NICHOLSON

R. JOHN STARMER

MICHAEL WEBER

EVERETT A. WICK

UNION OF CONCERNED SCIENTISTS

MICHAEL FADEN

UNITED KINGDOM DEPT OF THE

ENVIRONMENT

F. S. FEATES UNIVERSITY OF AKRON

LORETTA J. COLE

UNIVERSITY OF ARIZONA JAAK DAEMEN

JAMES G. MCCRAY UNIVERSITY OF CALIFORNIA AT BERKELEY

TODD LAPORTE

THOMAS H. PIGFORD

UNIVERSITY OF CALIFORNIA AT LOS ANGELES

D. OKRENT

UNIVERSITY OF FLORIDA DAVID E, CLARK

DOLORES C. JENKINS

M. I. OHANIAN UNIVERSITY OF HAWAII AT MANOA

DAVID EPP

UNIVERSITY OF MARYLAND MARVIN ROUSH

UNIVERSITY OF MICHIGAN

WILLIAM YERR

UNIVERSITY OF MINNESOTA

DONALD GILLIS UNIVERSITY OF MISSISSIPPI

GEORGE D. BRUNTON

UNIVERSITY OF MISSOURI AT ROLLA ARVIND KUMAR

NICK TSOULFANIDIS UNIVERSITY OF NEW MEXICO

HAROLD M. ANDERSON

DOUGLAS G. BROOKINS UNIVERSITY OF OKLAHOMA

UNIVERSITY OF OTTAWA - CANADA

DANIEL T. BOATRIGHT

TUNCER OREN UNIVERSITY OF PITTSBURGH

B. L. COHEN

UNIVERSITY OF SOUTHERN MISSISSIPPI CHARLES R. BRENT

JAMES W. PINSON UNIVERSITY OF TEXAS AT AUSTIN

BUREAU OF ECONOMIC GEOLOGY E. G. WERMUND

UNIVERSITY OF TEXAS AT SAN ANTONIO

DONALD R. LEWIS

UNIVERSITY OF TOKYO - JAPAN

RYOHEI KIYOSE

UNIVERSITY OF UTAH

JAMES W. BUNGER MARRIOTT LIBRARY

GARY M. SANDQUIST

UNIVERSITY OF WISCONSIN AT MILWAUKEE

HOWARD PINCUS

UPPER PEASE SOIL AND WATER

CONSERVATION DISTRICT

W.H. MARSHALL

URS/JOHN A. BLUME & ASSOCIATES,

ENGINEERS

ANDREW B. CUNNINGHAM

UTAH DEPT OF NATURAL RESOURCES &

ENERGY

MARK A. PAGE

UTAH DIVISION OF ENVIRONMENTAL

HEALTH

MARV H. MAXELL

UTAH DIVISION OF OIL, GAS & MINING

SALLY J. KEFER

UTAH DIVISION OF PARKS & RECREATION

JOHN KNUDSON

GORDON W. TOPHAM

UTAH ENERGY OFFICE

ROD MILLAR

UTAH ENVIPONMENT CENTER

JUNE WICKHAM

UTAH GEOLOGICAL AND MINERAL SURVEY

MAGE YONETANI

UTAH OFFICE OF PLANNING & BUDGET

RANDY MOON (25)

UTAH SCIENCE COUNCIL

RANDY MOON

UTAH SOUTHEASTERN DISTRICT HEALTH

DEPARTMENT

ROBERT L. FURLOW

UTAH STATE GEOLOGIC TASK FORCE

DAVID D. TILLSON

UTAHNS AGAINST THE DUMP COALITION

UTILITY DATA INSTITUTE

FRED YOST

VANDERBILT UNIVERSITY

FRANK L. PARKER

VERMONT DEPT OF WATER RESOURCES AND

ENVIRONMENTAL ENGINEERING

CHARLES A. RATTE

VERMONT STATE NUCLEAR ADVISORY PANEL

VIRGINIA CALLAN

VIRGINIA DEPT OF HEALTH

WILLIAM F. GILLEY

ROBERT G. WICKLINE

VIRGINIA DIVISION OF MINERAL RESOURCES

ROBERT C. MILICI

VIRGINIA HOUSE OF DELEGATES

A. VICTOR THOMAS

WASHINGTON HOUSE OF REPRESENTATIVES

RAY ISAACSON

WATTLAB BOB E. WATT

WEBSTER PARISH LIBRARY

WEST VALLEY NUCLEAR SERVICES CO INC

CHRIS CHAPMAN

ERICH J. MAYER

WESTINGHOUSE ELECTRIC CORP

GEORGE V. B. HALL

JAMES H. SALING

JAMES R. SCHORNHOUST

WOODWARD-CLYDE CONSULTANTS

F. R. CONWELL (2)

	,		

COMMENT SHEET

Comments		
	-	
<u> </u>		
		· · · · · · · · · · · · · · · · · · ·
		1.11
	(Use additional sheet if nece	ssary.)
Nama		Date
Organization		
Street		
Oit.	State	Zin Code

Fold Here	

MANAGER, DOE/NPO 505 KING AVENUE COLUMBUS, OHIO 43201 USA

#		

		•	
•			
		•	

a `		
	. 3	
•		
•		

BMI/ONWI-516 Sensitivity/Uncertainty Analysis of a Borehole Scenario Comparing Latin Hypercube Sampling and Deterministic Sensitivity Approaches



